

# An Introduction to SusyFit

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ERC Ideas: NPFlavour

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SAPIENZA  
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New Physics at Belle II. Karlsruhe. 25<sup>th</sup>Feb 2015.

**by far the most difficult part of the project has been giving the code  
a name... we still stand undecided...**

**SusyFit is how it started, now it is more than just susy.**

have suggestions? email me: [apaul2@alumni.nd.edu](mailto:apaul2@alumni.nd.edu)

# the program

- ✓ an analysis tool for direct and indirect observables
- ✓ comes with Bayesian Analysis Tool based on Markov Chain Monte Carlo
- ✓ SM and BSM arranged modularly for extraction of model based computations
- ✓ possibilities of adding user-generated models of new dynamics
- ✓ possibilities of adding user-defined observables
- ✓ possibilities of performing any choice of statistical analysis using the library
- ✓ a handy tool for getting very quick (statistical) estimates and doing full-fledged statistical analyses
- ✓ deployable both on clusters and multicore CPUs for large statistical analyses.
- ✓ equally friendly for all level of users and developers (doxygened in detail)

# the philosophy

everyone gets a candy they like

- we offer a variety of interfaces that can cater to beginners, advanced users and developers
- a variety of NP models and observables will be included and the developers can add more

statistical precision requires large samples

- a lot of focus has been put on speed with extensive caching built in
- built-in MPI parallelization for deployment on large clusters

open source and open for customization

- source will be released under GPL with extensive documentation
- working developer version always available through git (requires NetBeans IDE)

# the dependencies

ROOT (<https://root.cern.ch>)

- it is used to plot and store all the histograms generated at run-time (in \*.pdf, \*.ps and \*.root)
- for now we are using ROOT v5.xx
- ROOT v6.xx migration scheduled for later this year

BOOST (<http://www.boost.org/>)

- a largely header based implementation of efficient and safe memory accessing procedures and file parser in c++
- we are using the headers only so that building of boost is unnecessary

GSL (<http://www.gnu.org/software/gsl/>)

- the GNU Scientific Libraries: efficient matrix operations and integrals
- we have our own wrappers for GSL to aid future developers

# the optional dependencies

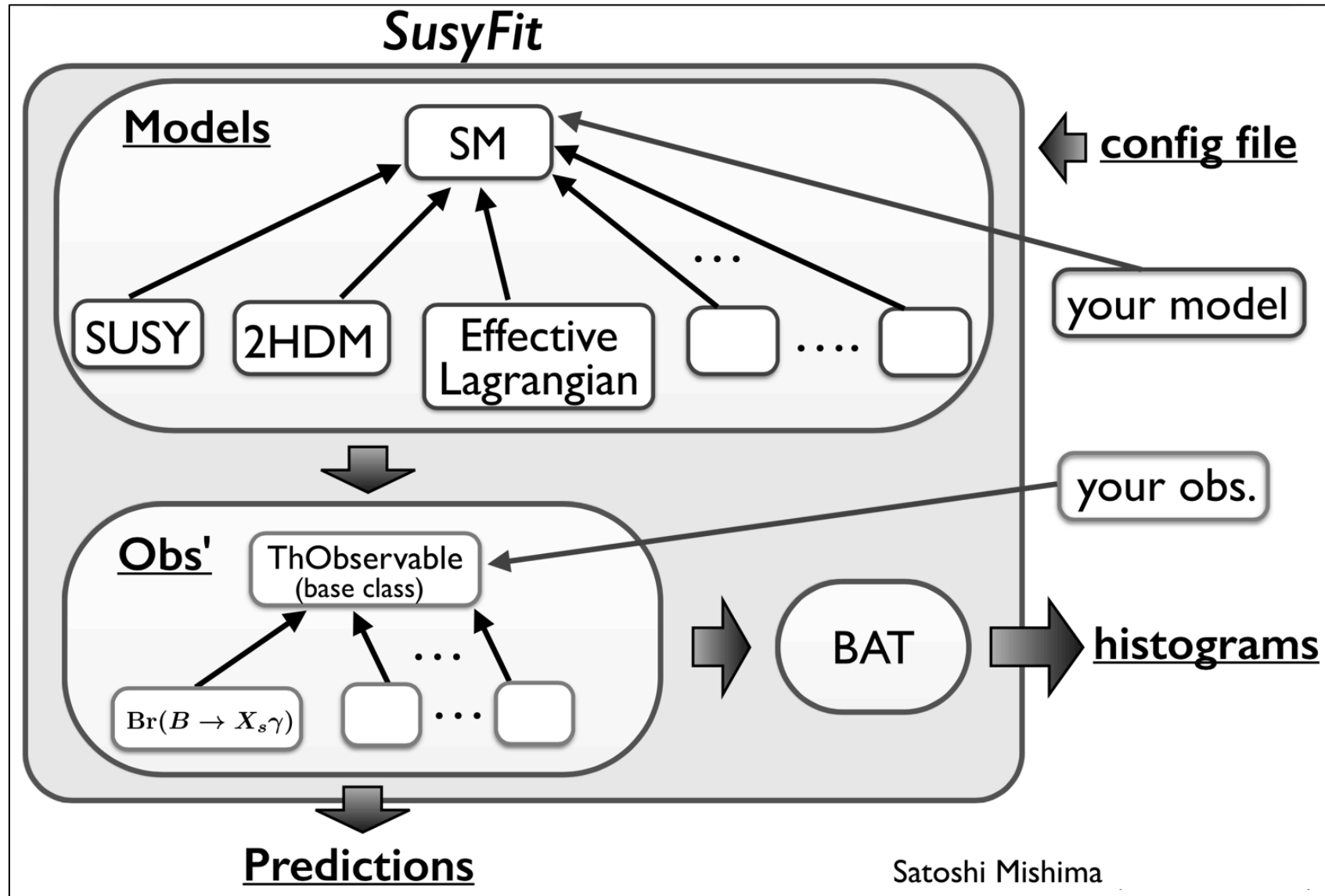
BAT (<https://www.mppmu.mpg.de/bat/>)

- necessary if our MCMC engine is used
- Bayesian Analysis Tool: developed separately as a code for Bayesian analysis based on Markov Chain Monte Carlo routines
- highly tunable in its procedures with different optimizing algorithms
- we provide access to some tuning parameters, others accessible through modification of our Monte Carlo engine
- currently compatible with BAT v0.9.3
- migration to BAT v0.9.4 scheduled for later this year

openMPI (<http://www.open-mpi.org/>) / MPICH (<http://www.mpich.org/>)

- necessary only for parallelized runs
- requires our patched version of BAT v0.9.3
- tested for large scale deployment @  $O(10^3)$  cores in batch submission systems

# the structure of the code



# the structure of the code

all model parameters set by the user through a configuration files

Model definition

```
StandardModel
#####
# Model Parameters
#          name          ave          errg          errf
#-----
### Parameters in StandardModel
ModelParameter GF          1.1663787e-5  0.          0.
### alpha=1/137.035999074
ModelParameter ale 7.2973525698e-3  0.          0.
ModelParameter AlsMz 0.1185 0.0005 0.
ModelParameter dAle5Mz 0.02750 0.00033 0.
ModelParameter Mz 91.1875 0.0021 0.
ModelParameter delMw 0. 0. 0.
ModelParameter delSin2th_l 0. 0. 0.
ModelParameter delGammaZ 0. 0. 0.
### mtpole
ModelParameter mtop 173.34 0.76 0.
ModelParameter mHl 125.5 0. 0.
#
### light quark masses at 2 GeV
ModelParameter mup 0.0023 0. 0.
ModelParameter mdown 0.0048 0. 0.
ModelParameter mstrange 0.0938 0.0024 0.
### mc(mc)
ModelParameter mcharm 1.3 0.03 0.
### mb(mb)
ModelParameter mbottom 4.18 0.03 0.
ModelParameter muc 1.3 0. 0.
ModelParameter mub 4.8 0. 0.
ModelParameter mut 164.1 0.07 0.
#
ModelParameter mneutrino_1 0. 0. 0.
ModelParameter mneutrino_2 0. 0. 0.
ModelParameter mneutrino_3 0. 0. 0.
ModelParameter melectron 5.109989e-4 0. 0.
ModelParameter mmu 0.10565837 0. 0.
ModelParameter mtau 1.77682 0. 0.
#####
# Mandatory configuration files
#-----
IncludeFile conf_files/Flavour.conf
```

Model parameters are set mandatory according to the model defined in the configuration file

additional mandatory/ optional configuration files



# the structure of the code

observables list set by user in the configuration files

```
#####  
# Observables  
# use one of the following formats:  
# Observable name th label min max (no)MCMC weight ave errg errf  
# Observable name th label min max (no)MCMC file filename histoname  
# Observable name th label min max noMCMC noweight  
#  
# BinnedObservables:  
# use one of the following formats:  
# BinnedObservable name th label min max (no)MCMC weight ave errg errf bin_min bin_max  
# BinnedObservable name th label min max (no)MCMC file filename histoname bin_min bin_max  
# BinnedObservable name th label min max noMCMC noweight bin_min bin_max  
#  
# Observables2D  
# use one of the following formats:  
# Observable2D name th1 label1 min1 max1 noMCMC noweight th2 label2 min2 max2  
# Observable2D name th1 label1 min1 max1 MCMC file filename histoname th2 label2 min2 max2  
#  
# The keyword "CorrelatedGaussianObservables name Nobs" initializes a set  
# of Nobs correlated observables. It must be followed by exactly Nobs  
# Observable lines and then by Nobs lines of Nobs numbers (the corr matrix).  
#-----  
#####
```

# the structure of the code

observables list set by user in the configuration files

```
#-----  
#####  
### 1304.6325  
BinnedObservable P_1_LQ1 P_1_BdKstmu P_1 -2.19 1.81 MCMC weight -0.19 0.40 0. 0.1 2.  
BinnedObservable P_1_LQ2 P_1_BdKstmu P_1 -3.54 2.96 MCMC weight -0.29 0.65 0. 2. 4.3  
BinnedObservable P_1_LQ3 P_1_BdKstmu P_1 -1.19 1.91 MCMC weight 0.36 0.31 0. 4.3 8.68  
BinnedObservable P_1_LQ4 P_1_BdKstmu P_1 -1.90 2.20 MCMC weight 0.15 0.41 0. 1. 6.  
#  
### 1304.6325  
BinnedObservable P_2_LQ1 P_2_BdKstmu P_2 -0.72 0.78 MCMC weight 0.03 0.15 0. 0.1 2.  
BinnedObservable P_2_LQ2 P_2_BdKstmu P_2 0.15 0.85 MCMC weight 0.50 0.07 0. 2. 4.3  
BinnedObservable P_2_LQ3 P_2_BdKstmu P_2 -0.65 0.15 MCMC weight -0.25 0.08 0. 4.3 8.68  
BinnedObservable P_2_LQ4 P_2_BdKstmu P_2 -0.27 0.93 MCMC weight 0.33 0.12 0. 1. 6.  
#  
### No Data  
BinnedObservable P_3_LQ1 P_3_BdKstmu P_3 1. -1. noMCMC noweight 1. 0. 0. 0.1 2.  
BinnedObservable P_3_LQ2 P_3_BdKstmu P_3 1. -1. noMCMC noweight 1. 0. 0. 2. 4.3  
BinnedObservable P_3_LQ3 P_3_BdKstmu P_3 1. -1. noMCMC noweight 1. 0. 0. 4.3 8.68  
BinnedObservable P_3_LQ4 P_3_BdKstmu P_3 1. -1. noMCMC noweight 1. 0. 0. 1. 6.  
#  
### 1308.1707  
BinnedObservable P_4p_LQ1 P_4p_BdKstmu P_4p -2.60 2.60 MCMC weight 0.00 0.52 0. 0.1 2.  
BinnedObservable P_4p_LQ2 P_4p_BdKstmu P_4p -2.26 3.74 MCMC weight 0.74 0.60 0. 2. 4.3  
BinnedObservable P_4p_LQ3 P_4p_BdKstmu P_4p -0.42 2.78 MCMC weight 1.18 0.32 0. 4.3 8.68  
BinnedObservable P_4p_LQ4 P_4p_BdKstmu P_4p -1.22 2.38 MCMC weight 0.58 0.36 0. 1. 6.  
#  
### 1308.1707  
BinnedObservable P_5p_LQ1 P_5p_BdKstmu P_5p -0.75 1.65 MCMC weight 0.45 0.24 0. 0.1 2.  
BinnedObservable P_5p_LQ2 P_5p_BdKstmu P_5p -1.71 2.29 MCMC weight 0.29 0.40 0. 2. 4.3  
BinnedObservable P_5p_LQ3 P_5p_BdKstmu P_5p -0.99 0.61 MCMC weight -0.19 0.16 0. 4.3 8.68  
BinnedObservable P_5p_LQ4 P_5p_BdKstmu P_5p -0.84 1.26 MCMC weight 0.21 0.21 0. 1. 6.  
#
```

# the structure of the code

Parametric correlations set through the CorrelatedGaussianObservables method

```
#####  
CorrelatedGaussianObservables LatticeV 2  
Observable a_0V a_0V a_0V 1 -1 MCMC weight 0.4975 0.0667 0.  
Observable a_1V a_1V a_1V 1 -1 MCMC weight -2.0151 0.9165 0.  
1. 0.859005  
0.859005 1.  
#  
CorrelatedGaussianObservables LatticeA0_A12 4  
Observable a_0A0 a_0A0 a_0A0 1 -1 MCMC weight 0.5023 0.0370 0.  
Observable a_1A0 a_1A0 a_1A0 1 -1 MCMC weight -1.6084 0.4473 0.  
Observable a_0A12 a_0A12 a_0A12 1 -1 MCMC weight 0.2196 0.0238 0.  
Observable a_1A12 a_1A12 a_1A12 1 -1 MCMC weight 0.3324 0.3002 0.  
1. 0.667316 0.904271 0.887787  
0.667316 1. 0.909064 0.927202  
0.904271 0.909064 1. 0.97564  
0.887787 0.927202 0.97564 1.  
#  
CorrelatedGaussianObservables LatticeA1 2  
Observable a_0A1 a_0A1 a_0A1 1 -1 MCMC weight 0.2848 0.0233 0.  
Observable a_1A1 a_1A1 a_1A1 1 -1 MCMC weight 0.1914 0.2804 0.  
1. 0.948261  
0.948261 1.  
#  
CorrelatedGaussianObservables LatticeT1_T2 4  
Observable a_0T1 a_0T1 a_0T1 1 -1 MCMC weight 0.4197 0.0241 0.  
Observable a_1T1 a_1T1 a_1T1 1 -1 MCMC weight -1.3633 0.2586 0.  
Observable a_0T2 a_0T2 a_0T2 1 -1 MCMC weight 0.27997 0.01948 0.  
Observable a_1T2 a_1T2 a_1T2 1 -1 MCMC weight 0.1171 0.2364 0.  
1. 0.500481 0.850285 0.820455  
0.500481 1. 0.801509 0.836394  
0.850285 0.801509 1. 0.93236  
0.820455 0.836394 0.93236 1.  
#  
CorrelatedGaussianObservables LatticeT23 2  
Observable a_0T23 a_0T23 a_0T23 1 -1 MCMC weight 0.5235 0.0451 0.  
Observable a_1T23 a_1T23 a_1T23 1 -1 MCMC weight -0.2714 0.5791 0.  
1. 0.951964  
0.951964 1.  
#
```

# the structure of the code

all MCMC parameters set through a separate configuration file.

```
## Number of chains
NChains          24

## Max iterations in prerun
PrerunMaxIter    200000

## Analysis iterations
Iterations       200000

## Write Markov Chain
WriteChain       false

## use a particular seed
Seed            0

## BAT tuning parameters
FindModeWithMinuit  false
CalculateEvidence   false
PrintAllMarginalized true
PrintCorrelationMatrix false
PrintKnowledgeUpdatePlots false
PrintParameterPlot  false
OrderParameters    true
```

# sample user codes for MCMC

```
#include <iostream>
#include <SusyFit.h> ← the header

#ifdef _MPI
#include <mpi.h>
#endif

int main(int argc, char** argv)
{
#ifdef _MPI
    MPI::Init();
    int rank = MPI::COMM_WORLD.Get_rank();
    MPI::Status status;
#else
    int rank = 0;
#endif

    try {

        if(argc != 3){
            if (rank == 0) std::cout << "\nusage: " << argv[0] << " ModelConf.conf MonteCarlo.conf\n" << std::endl;
            return EXIT_SUCCESS;
        }

        std::string ModelConf = argv[1];
        std::string MCMCConf = argv[2];
        std::string FileOut = "";
        std::string JobTag = "";

        {
            ThObsFactory ThObsF;
            ModelFactory ModelF;
            MonteCarlo MC(ModelF, ThObsF, ModelConf, MCMCConf, FileOut, JobTag);
        } ← the user code
        MC.Run(rank);

#ifdef _MPI
        MPI::Finalize();
#endif

        return EXIT_SUCCESS;
    } catch (const std::runtime_error& e) {
        std::cerr << e.what() << std::endl;
        return EXIT_FAILURE;
    }
}
```

# to implement your own statistical analysis

```
#include <iostream>
#include <ComputeObservables.h>

int main(int argc, char** argv)
{
    try {
        std::string ModelConf = argv[1];
        std::map<std::string, double> DPars;

        ThObsFactory ThObsF;
        ModelFactory ModelF;

        ComputeObservables CO(ModelF, ThObsF, ModelConf);

        {
            CO.AddObservable("Mw");
            CO.AddObservable("GammaZ");
            CO.AddObservable("AFBbottom");
        } ← list observables

        std::map<std::string, double> DObs = CO.getObservables();

        for (int i = 0; i < 2; i++) {

            DPars["Mz"] = 91.1875 + 0.0001 * i;
            DPars["AlsMz"] = 0.1184 + 0.000001 * i;

            DObs = CO.compute(DPars);

            std::cout << "\nParameters[" << i + 1 << "]:" << std::endl;
            for (std::map<std::string, double>::iterator it = DPars.begin(); it != DPars.end(); it++) {
                std::cout << it->first << " = " << it->second << std::endl;
            }
            std::cout << "\nObservables[" << i + 1 << "]:" << std::endl;
            for (std::map<std::string, double>::iterator it = DObs.begin(); it != DObs.end(); it++) {
                std::cout << it->first << " = " << it->second << std::endl;
            }
        }

        return EXIT_SUCCESS;
    } catch (const std::runtime_error& e) {
        std::cerr << e.what() << std::endl;
        return EXIT_FAILURE;
    }
}
```

your own statistical analysis

# an option for the hardcore

```
#include <iostream>
#include <ComputeObservables.h>
#include <InputParameters.h>
```

← Header containing all mandatory input parameters

```
int main(int argc, char** argv)
{
```

```
  try {
```

```
    std::string ModelName = "NPEpsilons";
```

```
    {
      InputParameters IP;
      std::map<std::string, double> DPars_IN = IP.getInputParameters(ModelName);
      DPars_IN["mcharm"] = 1.3;
      DPars_IN["mub"] = 4.2;
    }
```

← do it yourself...

```
    ComputeObservables CO(ModelName, DPars_IN);
```

```
    CO.AddObservable("Mw");
    CO.AddObservable("GammaZ");
    CO.AddObservable("AFBbottom");
```

```
    std::map<std::string, double> DObs = CO.getObservables();
```

```
    std::map<std::string, double> DPars;
```

```
    for (int i = 0; i < 2; i++) {
```

```
      DPars["mtop"] = 170.0 + i * 0.1;
      DPars["dAle5Mz"] = 0.02750 - i * 0.0001;
```

```
      DObs = CO.compute(DPars);
```

```
      std::cout << "\nParameters[" << i + 1 << "]:" << std::endl;
```

```
      for (std::map<std::string, double>::iterator it = DPars.begin(); it != DPars.end(); it++) {
        std::cout << it->first << " = " << it->second << std::endl;
      }
```

```
      std::cout << "\nObservables[" << i + 1 << "]:" << std::endl;
```

```
      for (std::map<std::string, double>::iterator it = DObs.begin(); it != DObs.end(); it++) {
        std::cout << it->first << " = " << it->second << std::endl;
      }
```

```
    }
```

```
    return EXIT_SUCCESS;
```

```
  } catch (const std::runtime_error& e) {
    std::cerr << e.what() << std::endl;
    return EXIT_FAILURE;
  }
```

```
}
```

↑  
your own statistical analysis

# nail-biting time?

- observable computation rate of  $O(\text{kHz})$  or better for the implemented observables
- a simple test analysis can be done in minutes (statistics of  $O(10\text{k})$ )
- a results analysis can be done in hours (statistics of  $O(100\text{k})$ )
- a publication level analysis can be done in days (statistics of  $O(1\text{M})$  or better)
- times are for single thread (core) single chain analyses on a laptop/desktop
- using the built in MPI implementation reduces time or increases statistics by a factor of  $O(N)$  for  $N$  threads (cores) used

caveat: computers only get faster...

➡ made a mistake? the code exits with a message telling you why.

caveat: no code is seg fault proof, however, error handling is one of our priorities



# model menu

- **Standard Model** (fully tested and results available in the literature with more to come soon)
- **general MSSM** (including variants like MFV, pMSSM etc.) with SLHA2 compatibility  
**FeynHiggs** to be used as the spectrum generator
- **two Higgs doublet models** (under construction)
- **some model independent extensions** for the study of NP in EW and Higgs physics (dim-6 operators, oblique parameters etc. tested)

to write a custom model one can use the template that we will provide, adding the necessary input and derived parameters

custom observables can also be added which can depend on the parameters of the existing models and/or the custom model

# observables menu

- **EW precision observables** (tested)

$$M_W, \Gamma_W, \Gamma_Z, \sigma_h^0, \sin^2 \theta_{\text{eff}}^{\text{lept}} (Q_{\text{FB}}^{\text{had}}), P_{\tau}^{\text{pol}}, \mathcal{A}_f, A_{\text{FB}}^{0,f}, R_f^0$$

for  $f = \ell, c, b$

- **Higgs boson signal strengths** (tested)

$$H \rightarrow \gamma\gamma, ZZ, WW, \tau^+\tau^-, b\bar{b} \text{ for different categories}$$

- **Lep2 two fermion processes** (tested)

$$\sigma \text{ and } A_{\text{FB}} \text{ for } e^+e^- \rightarrow e^+e^-, \mu^+\mu^-, \tau^+\tau^-, c\bar{c}, b\bar{b}$$

# observables menu

- **Unitarity triangle observables** (tested against UFit)

UT angles,  $\Delta F = 2$  amplitudes, CKM elements

- **rare decays** (under development)

$$B \rightarrow X_s \gamma, B \rightarrow K^* \gamma \quad (\text{in progress})$$

$$B \rightarrow X_s \ell^+ \ell^-, B \rightarrow K \ell^+ \ell^- \quad (\text{in progress})$$

$$B \rightarrow K^* \ell^+ \ell^-$$

$$B_{s,d} \rightarrow \mu^+ \mu^-$$

$$K \rightarrow \pi \nu \bar{\nu} \quad (\text{in progress})$$

$$K \rightarrow \mu^+ \mu^- \quad (\text{in progress})$$

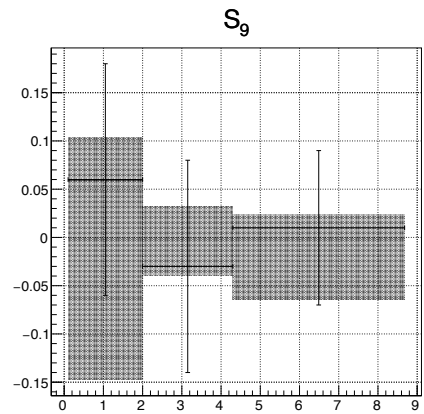
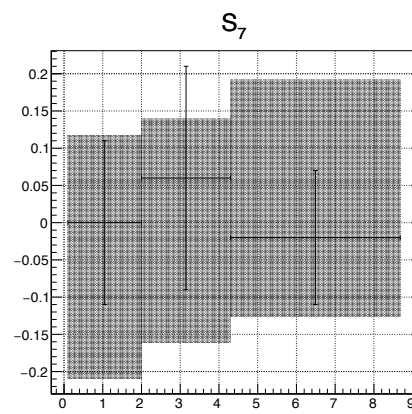
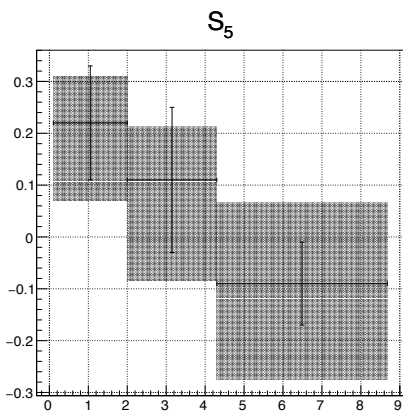
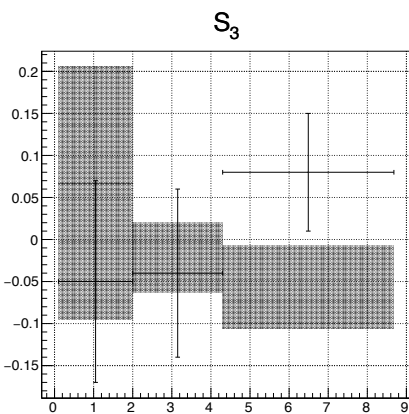
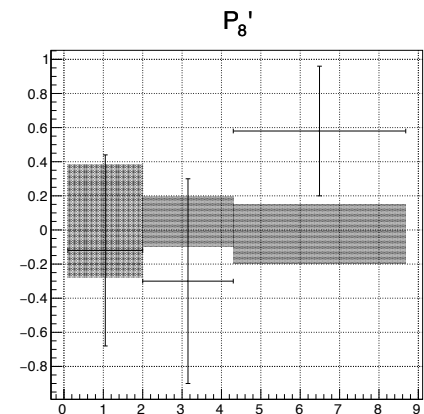
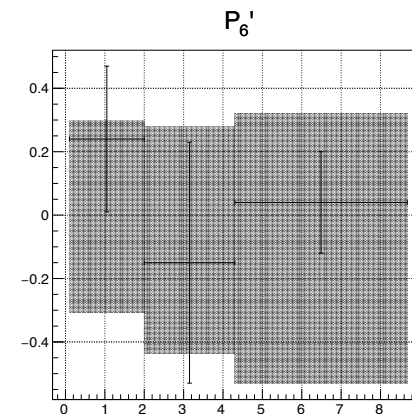
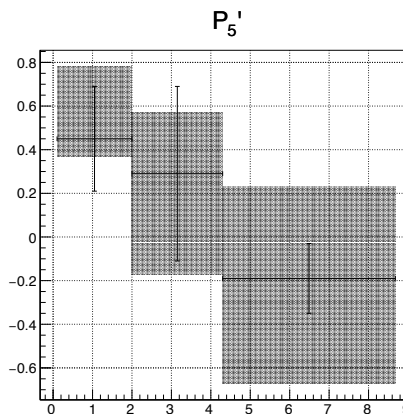
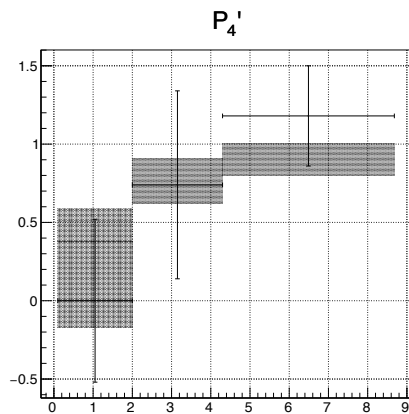
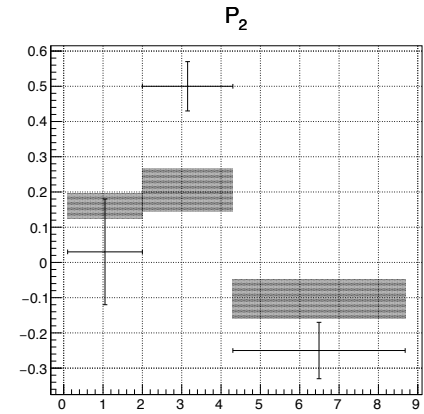
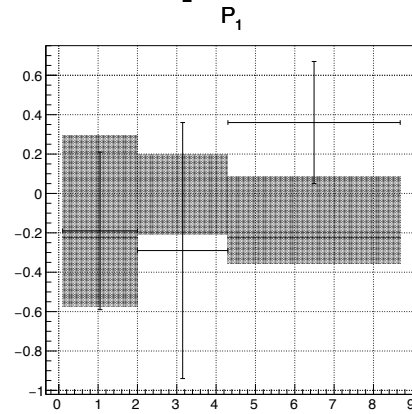
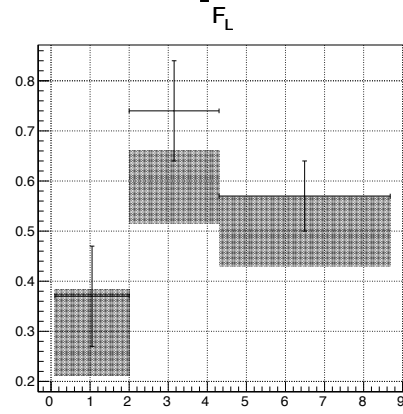
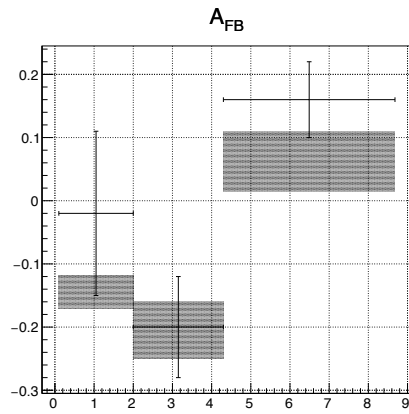
$$\tau \rightarrow \mu \gamma, \tau \rightarrow 3\ell \quad (\text{+other LFV processes, in progress})$$

- **non-leptonic decays** (tested)

$$B \rightarrow PP, PV \quad (\text{in progress})$$

$$\epsilon'/\epsilon \quad (\text{in progress})$$

# proof of concept



# utility and prospects

- **theory predictions:**

this framework can simply be used for theory predictions of observables for a certain set or a range of the set of parameters

- **phenomenological studies:**

this can also be used for detailed phenomenological studies under a chosen statistical framework

- **correlations:**

this can be used for studies of correlations between different observables within a given parameter space of a given model

- **theoretical uncertainties:**

a very effective tool for deriving parametric uncertainties coming from the whole or part of the theory parameter space

## summary

- ✓ we are developing a computational framework that we hope will be of much utility to both theorist and experimentalists
- ✓ a tool that will combine direct and indirect searches
- ✓ it will allow for easy comparisons and phenomenological analyses of experimental results in the light of theoretical frameworks within the Standard Model and beyond
- ✓ a flexible framework that allows for different kinds of statistical analyses with a Bayesian analysis built in based on very efficient and fast Markov Chain Monte Carlo routines
- ✓ a modular structure that allows for users to modify or add to the existing one
- ✓ uses the leading industry standards in MPI, GSL, Boost and ROOT
- ✓ will come with full documentation in the form of code documentation (doxygen) and an “instruction manual” for users including physics references

# the usual debate...

DID THE SUN JUST EXPLODE?  
(IT'S NIGHT, SO WE'RE NOT SURE.)

THIS NEUTRINO DETECTOR MEASURES  
WHETHER THE SUN HAS GONE NOVA.

THEN, IT ROLLS TWO DICE. IF THEY  
BOTH COME UP SIX, IT LIES TO US.  
OTHERWISE, IT TELLS THE TRUTH.

LET'S TRY.

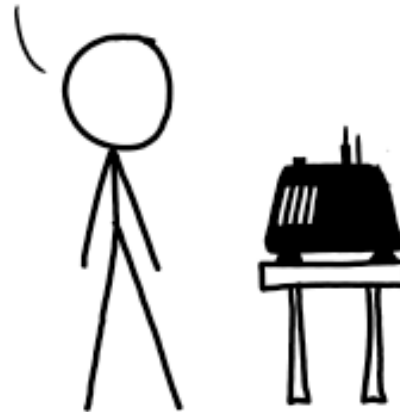
DETECTOR! HAS THE  
SUN GONE NOVA?

ROLL  
YES.



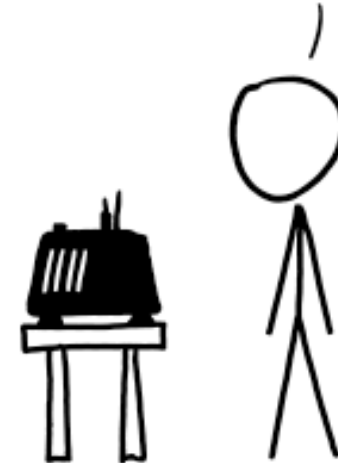
FREQUENTIST STATISTICIAN:

THE PROBABILITY OF THIS RESULT  
HAPPENING BY CHANCE IS  $\frac{1}{36} = 0.027$ .  
SINCE  $p < 0.05$ , I CONCLUDE  
THAT THE SUN HAS EXPLODED.



BAYESIAN STATISTICIAN:

BET YOU \$50  
IT HASN'T.



now you have options...

# the developers

## ***Roma:***

Shehu S. AbdusSalam

Jorge de Blas

Debtosh Chowdhury

Otto Eberhardt

Marco Fedele

Enrico Franco

Diptimoy Ghosh

Ayan Paul

Luca Silvestrini

## ***Roma Tre:***

Marco Ciuchini

## ***KIAS:***

Satoshi Mishima

## ***ICTP/SISSA:***

Giovanni Grilli di Cortona

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*We look forward to welcoming developers willing to add models/observables to the existing framework and/or run tests against codes existing in the market.*



**in a few decades we will put an artificial intelligence engine in the code...**

**Thank you...!!**



To my Mother and Father, who showed me what I could do,  
and to Ikaros, who showed me what I could not.

“To know what no one else does, what a pleasure it can be!”

– adopted from the words of  
Eugene Wigner.

